

Q: Why does EM work?

Let $f(y, z; \theta)$ be the density for the complete data (y, z)

Recall $f(y; \theta) = \int f(y, z; \theta) dz$ is the marginal density of Y .

log-likelihood: $l(\theta) = \ln f(y; \theta)$

Let θ^v be the current value - calculate difference $l(\theta) - l(\theta^v)$

$$l(\theta) - l(\theta^v) = \ln f(y; \theta) - \ln f(y; \theta^v)$$

$$= \ln \int f(y, z; \theta) dz - \ln f(y; \theta^v)$$

$$= \ln \int \frac{f(y, z; \theta)}{f(z|y; \theta^v)} \cdot \frac{f(z|y; \theta^v)}{f(z|y; \theta^v)} dz - \ln f(y; \theta^v)$$

multiply/divide by conditional density $z|y$ when $\theta = \theta^v$

$$= \ln \int \frac{f(y, z; \theta)}{f(z|y; \theta^v)} f(z|y; \theta^v) dz - \ln f(y; \theta^v)$$

$$= \ln E_{z|y; \theta^v} \left[\frac{f(y, z; \theta)}{f(z|y; \theta^v)} \right] - \ln f(y; \theta^v)$$

[ln is "concave" for \Rightarrow Jensen's inequality $\ln EX \geq E \ln X$]

$$\geq E_{z|y; \theta^v} \left(\ln \frac{f(y, z; \theta)}{f(z|y; \theta^v)} \cdot \frac{1}{f(y; \theta^v)} \right)$$

$$= E_{z|y; \theta^v} \left[\ln \frac{f(y, z; \theta)}{f(y, z; \theta^v)} \right]$$

In short

$$l(\theta) - l(\theta^v) \geq E_{Z|Y; \theta^v} \left[\ln \frac{f(Y, Z; \theta)}{f(Y, Z; \theta^v)} \right] = Q(\theta; \theta^v, Y)$$

when $\theta = \theta^v \Rightarrow \text{RHS} = 0$.

Max RHS
is equivalent
to maximizing
 $Q(\theta; \theta^v, Y)$

Thus if we choose $\theta = \theta^{v+1}$ that maximizes the RHS it is fair to say that

$$E_{Z|Y; \theta^v} \left[\ln \frac{f(Y, Z; \theta^{v+1})}{f(Y, Z; \theta^v)} \right] \geq 0$$

$$\Rightarrow l(\theta^{v+1}) - l(\theta^v) \geq 0$$

Note that maximizing the RHS wrt θ really means maximizing

$$E_{Z|Y; \theta^v} \left[\ln f(Y, Z; \theta) \right] \text{ wrt } \theta$$

This is exactly the "Q"-function.

Choosing the updates for θ in this way guarantees that likelihood increases.

ged.

Last remarks

θ^v may be a stationary point which need not be local max
In some cases EM converges to a local minimum or a saddlepoint
However if the likelihood is unimodal and satisfies some differentiability conditions, then EM converges to the unique MLE.